



PROCESS ALLOWING IMMIGRATION, II: THE

CASE OF INFINITE OFFSPRING MEAN

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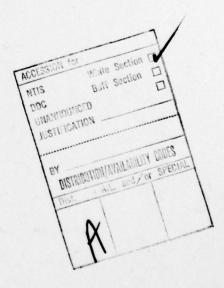
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## ABSTRACT

This paper obtains some limit theorems for the simple branching process allowing immigration  $\{X_n\}$  when the offspring mean is infinite. It is shown that there exists a function U such that  $\{e^{-n}U(X_n)\}$  converges almost surely and if  $s = \Sigma b_j \log^+ U(j) < --$ , where  $\{b_j\}$  is the immigration distribution, the limit is non-defective and non-degenerate but is infinite if s = --.

When s = -1 limit theorems are found for  $\{U(X_n)\}$  which involve a slowly varying non-linear norming.

Key words: Bienaymé-Galton-Watson branching process, immigration, martingale convergence, limit theorems, regular variation.



## 1. INTRODUCTION

Let  $\{X_n: n=0,1,\ldots\}$  denote a Bienaymé-Galton-Watson process with immigration, for which the offspring probability generating function, f, satisfies m=f'(1-)=- and  $\{I_n: n=1,2,\ldots\}$ , the number of immigrants into successive generations, are independent and identically distributed with probability generating function b satisfying b(0)<1. Suppose also that  $X_0=0$ .

This paper describes the behavior of  $X_n$  as n+- in terms of appropriate limit theorems. In §2 we show that for a certain increasing function U constructed as in [6], the sequence  $\{e^{-n}U(X_n)\}$  converges almost surely and that the limit is non-defective if a certain condition on the immigration distribution obtains and is essentially infinite otherwise. It is pointed out that, in the former case,  $\{X_n\}$  may be classified in terms of regularity and irregularity exactly as in the case of no immigration [6].

In §3 we consider the case where  $Y_n = e^{-n}U(X_n) + -a.s.$  Limit theorems are obtained for  $Y_n$  which involve a non-linear norming by a slowly varying (SV) function. Theorem 2 involves no extra assumptions but has a rather strange appearance: neater versions are given in Corollaries 1-3 under conditions expressed in terms of  $G(x) = 1 - b(\exp[-1/V(e^X)])$ , where V is the functional inverse of U. These results are analogues of the families of limit theorems for the case m < -a which were presented in Part I [3], and some examples are given to show that the conditions in Corollaries 1-3 are not vacuous. Finally, it is shown that there is no sequence  $(c_n)$  of norming constants such that  $Y_n/c_n$  has a limit in distribution which is neither defective nor degenerate at the origin.

To complete the survey in Part I, it suffices to mention that a version of Theorem 1 was first proved in [2] under conditions on f ensuring that U could be chosen as a suitable power of log<sup>†</sup> and only convergence in law was established. The conditions of this result were substantially relaxed in [1] to allow U to be SV at infinity and strictly increasing. A restricted and more opaque version of Corollary 1 was proved in [2].

## 2. ALMOST SURE CONVERGENCE

Let  $f_n$  denote the  $n^{th}$  functional iterate of f and  $h_n$  the functional inverse of  $-\log f_n(e^{-s})$ . Let  $p_n(s) = \prod_{m=1}^n b \left( \exp(-h_m(s)) \right)$  and  $p(s) = \lim_{n \to \infty} p_n(s)$  for  $0 < s < -\log q$  where q = f(q) < 1. Since  $h_n(s) \downarrow 0$   $(n \leftrightarrow \infty)$  it is clear that either p(s) > 0 or  $\equiv 0$ . In the former case it is clear that  $\{[\exp(-h_n(s)X_n)]/p_n(s)\}$  is a martingale and the martingale convergence theorem shows that for each a.s.  $0 < s < -\log q$ ,  $\exp(-h_n(s)X_n) \to W(s)$ , say, and E(s) = p(s). The behavior of  $\{X_n\}$  can be analyzed in precisely the same way in which that of the corresponding process without immigration was analyzed in [6]. In particular the classification of points in  $(0, -\log q)$  as regular and irregular carries over to the present case and if  $T = \sup\{s \mid 0 < s < -\log q$ ,  $W(s) < 1\}$  it follows that T is non-defective and  $P(T \ge s_p) = p(s_p)$ ,  $P(T = s_p) = 0$  for every regular point  $s_p$ .

Finally, for the function U constructed in [6], the arguments used in this reference show that  $e^{-n}U(X_n) \xrightarrow{a.s.} U(T^{-1})$ . In the complementary case, when  $p(s) \equiv 0$ , the martingale convergence theorem shows that  $\exp(-h_n(s)X_n) \xrightarrow{a.s.} 0$  for each  $0 < s < -\log q$ . To proceed

further we need the following details about U. It is shown in [6] that by starting with a fixed  $s_0 \in (0, -\log q)$  it is possible to construct U so that it is continuous on  $[0, \infty)$ , identically zero on  $[0, 1/(-\log q)]$ , positive and strictly increasing to infinity on  $[1/(-\log q), \infty)$  and satisfies the relations

$$U(1/h_n(s_0)) = e^n, U(1/h_n(s))/U(1/h_n(s_0)) = U(1/s)$$
 (1)

Choose a sequence  $\{s_m: m=1,2,\ldots\}$  such that  $-\log q > s_m + 0 \pmod{m+\infty}$ . Let  $\Omega_m$  be a subset of the basic probability space such that  $P(\Omega_m) = 1$  and  $h_n(s_m)X_n(\omega) + \infty \pmod{m+\infty}$ . Thus on  $\Omega' = \bigcap_{m \geq 1} \Omega_m$ ,  $h_n(s_m)X_n(\omega) + \infty$  for each m and this implies that eventually  $X_n(\omega) \geq 1/h_n(s_m)$  and hence, from (1), eventually  $e^{-n}U(X_n(\omega)) \geq U(1/s_m)$   $(\omega \in \Omega')$ .

Letting  $n \leftrightarrow \infty$  and them  $m \leftrightarrow \infty$ , we obtain all except the final assertion of the following theorem.

Theorem 1. If p(s) > 0 then  $e^{-n}U(X_n) \xrightarrow{a.s.} U(1/T)$  where T is defined above, and if p(s) = 0,  $e^{-n}U(X_n) \xrightarrow{a.s.} -$ . Furtheremore, p(s) > 0 iff

$$\sum_{j=1}^{n} b_{j} \log^{+} U(j) < \infty$$
 (2)

where {b<sub>i</sub>} denotes the immigration distribution.

The last assertion follows from these observations. Since U is strictly increasing, the relations (1) can be solved to obtain an explicit expression

$$h_n(s) = 1/V(e^nU(s^{-1}))$$
 (3)

for  $h_n$ , and the proof of Lemma 2 in [1] needs, with one exception,

only trivial changes to show that (2) holds iff

$$\prod_{n=1}^{\infty} b\left(\exp\left(-1/V(e^nU(s^{-1}))\right)\right) > 0$$

where V is the functional inverse of U, whence the assertion in this case. The exception is that the proof given in [1] assumes that log U is SV: that this is so follows from

Lemma 1. If L is SV at infinity then so is L(U(·)).

<u>Proof.</u> By virtue of the uniform convergence theorem for SV functions [7] it suffices to show that for each  $\lambda > 0$ ,  $U(\lambda x)/U(x)$  is bounded away from zero and infinity for all sufficiently large x. It suffices to consider the case  $1 \le \lambda < \infty$ . The proof to be given parallels that of Lemma 2.3.3 in [6]. Choose  $s_0 \in (0, -\log q)$  as in Construction 2.3.1 of [6], and set  $D = [h(s_0), s_0]$ . The functions  $g_n(s) = h_{n+1}(s)/h_n(s)$  are continuous on D, and, from equations (2) and (3) of [6], it follows that

$$g_n(s) \ge g_{n+1}(s)$$
;  $g_n(s) \to 0$  as  $n \to \infty$ : (4)

hence, by Dini's theorem, this convergence is uniform on D. Thus for all  $n \ge n_0$ , chosen large enough, and all  $s \in D$ ,  $\lambda/h_n(s) \le 1/h_{n+1}(s)$ . For each sufficiently large x construct n and  $s \in D$  so that  $x^{-1} = h_n(s)$  (see Construction 2.3.1, [5]). It follows that

$$1 < U(\lambda x)/U(x) \le U(1/h_{n+1}(s))/U(1/h_n(s)) = e$$
,

and the lemma now follows.

## 3. THE CASE p(s) = 0

The description of the behavior of  $X_n$  in the case  $p(s) \equiv 0$  is through distributional, rather than almost sure, limit theorems.

The first step in deriving them is to note that, from (3),

$$E\left(\exp\left(-h_n(t)X_n\right)\right) = p_n(t) = \prod_{m=1}^n b\left[\exp\left(-1/V\left(e^mU(t^{-1})\right)\right)\right].$$

Using integral test comparisons, as in [3], it is not difficult to show that

$$p_n(t) = \Delta_n(t) \exp \int_0^n \log b \left[ \exp \left( -1/V \left( e^y U(t^{-1}) \right) \right) \right] dy$$

where  $\Delta_n(t_n)+1$  if  $t_n+0+$  as  $n+\infty$ . Denoting the integral by  $J_n$  and making the change of variable  $e^X=e^YU(t^{-1})$ , it follows that if  $t=t_n+0+$  then

$$-J_n \sim J_n = \int_{\log U(1/t_n)}^{n+\log U(1/t_n)} G(x) dx$$
 (5)

where  $G(x) = 1 - b \left( exp(-1/V(e^X)) \right)$ . We are now in a position to approach the main limit theorem, by making a suitable choice of  $t_n$ , and by using the following version of Reuter's Lemma 1 in [1].

Lemma 2. Let U be constructed according to Construction 2.3.1 in

[6] and in addition suppose it is strictly increasing on  $(1/(-\log q), \infty)$ Let  $y_n$  be positive, increasing and satisfy the properties

(a) 
$$y_n(u)/y_n(v)+\infty$$
  $(n+\infty)$  if  $u>v$  and  $0,  $v< d \le \infty$ ; and$ 

(b) 
$$e^n y_n(u) \rightarrow \infty$$
  $(n \rightarrow \infty; c < u < d)$ .

If, for a sequence of non-negative random variables {W<sub>n</sub>} there is a continuous function a(·) such that

$$E\left[\exp\left(-W_{n}/V\left(e^{n}y_{n}(u)\right)\right)\right] \rightarrow a(u) \qquad (c < u < d; n+\infty)$$

then

$$P\left(e^{-n}U(W_n) \leq y_n(u)\right) + a(u).$$

<u>Proof.</u> Let  $A_n = \{e^{-n}U(W_n) \le y_n(u)\} = \{W_n \le V(e^ny_n(u))\}$ . Choose  $u\varepsilon(c,d)$  and  $c < u_1 < u < u_2 < d$ . If  $Y_n^{(i)} = \exp\{-W_n/V(e^ny_n(u_i))\}$  (i=1,2); then arguing as in [1] we obtain

$$EY_n^{(1)} - exp(-\lambda_n^{(1)}) \le P(A_n) \le (EY_n^{(2)}) exp \lambda_n^{(2)}$$

where  $\lambda_n^{(i)} = V(e^n y_n(u))/V(e^n y_n(u_i))$  (i=1,2). If we can show that  $\lambda_n^{(1)} + \infty$  and  $\lambda_n^{(2)} + 0$ , the assertion follows upon then letting  $u_i + u$ . We shall prove only that  $\lambda_n^{(1)} + \infty$ . It follows from (a) that for each t > 1

$$\lambda_{n}^{(1)} \ge V(te^{n}y_{n}(u_{1}))/V(e^{n}y_{n}(u_{1}))$$

for all sufficiently large n. Let m(n) be the integer part of  $n + \log y_n(u_1)$ . By (b)  $m(n) \rightarrow \infty$   $(n \rightarrow \infty)$  and  $e^{m(n)} \le e^n y_n(u_1) \le e^{m(n)+1}$  and hence, taking  $t = e^2$ ,

 $\lambda_n^{(1)} \ge V(\exp\{m(n)+2\})/V\left(\exp\{m(n)+1\}\right) = h_{m(n)}(1/V(e))/h_{m(n)+1}(1/V(e)),$  where we have used (3). Hence, from (4),  $\lim_{n \to \infty} \lambda_n^{(1)} = \infty$ .

Now, for  $x \ge 1$ , let  $\Lambda(x) = \exp \int_0^{\log x} \frac{n+\infty}{G(y)dy}$ , which is SV at

infinity and strictly increasing. Furthermore, the ratio

$$\Lambda(x)/\Lambda(xe^n) = \exp\left\{-\int_1^{\log x} G(y)dy\right\}$$
$$= \exp\left(-\int_1^{e^n} G(\log(zx))dz/z\right)$$

increases with x on  $(1,\infty)$  from  $m_n$  say, to unity, and since  $p(t) \equiv 0$  iff  $\int_0^\infty G(x) dx = \infty$ , it follows that  $m_n + 0$   $(n + \infty)$ ; moreover, for each fixed  $x \geq 1$ ,  $\Lambda(x)/\Lambda(xe^n)$  decreases to zero as  $n + \infty$ . Hence, for any 0 < u < 1 and all n sufficiently large, we may define  $y_n(u) > 0$  uniquely in such a way that

$$\Lambda(y_n(u)) = u\Lambda(e^ny_n(u))$$
.

where it follows that  $y_n$  is increasing on (0,1), that  $y_n(u)+\infty$   $(n+\infty)$  and that if 0 < v < u < 1,

$$\frac{y_n(u)}{y_n(v)} = \frac{\Lambda^{-1}\left(u\Lambda\left(e^ny_n(u)\right)\right)}{\Lambda^{-1}\left(v\Lambda\left(e^ny_n(v)\right)\right)} \ge \frac{\Lambda^{-1}\left(u\Lambda\left(e^ny_n(v)\right)\right)}{\Lambda^{-1}\left(v\Lambda\left(e^ny_n(v)\right)\right)} \longrightarrow$$

as  $n+\infty$ , since  $\Lambda$  is SV at infinity. Choosing  $U(t_n^{-1})=y_n(u)$  in (5) we see that  $j_n=u$  and hence  $\mathbb{E}\Big[\exp\Big(-X_n/V\big(e^ny_n(u)\big)\Big]\Big]+u$  (0<u<1), and that the other conditions of Lemma 2 are satisfied. This completes the proof of the following theorem.

Thereom 2. If 
$$p(s) \equiv 0$$
, then
$$\Lambda(e^{-n}U(X_n))/\Lambda(U(X_n)) \stackrel{d}{\rightarrow} W$$

where W is uniformly distributed on [0,1].

Analogues of Theorem 2 also obtain when m < 1 and 1 < m < m; see [3].

The proof of Theorem 2 has been carried out under the condition  $X_0 = 0$ . However

$$P_{i}^{(n)}(t) = E\left[\exp(-x_{n}h_{n}(t))|x_{0}=i\right]$$

$$= \left[f_{n}\left(\exp(-h_{n}(t))\right]^{i}P_{0}^{(n)}(t)$$

$$= e^{-it}P_{0}^{(n)}(t),$$

and since t was chosen to converge to zero, we see that Theorem 2 is valid for any initial state. Furthermore,  $\Lambda(e^{-n}U(\cdot))/\Lambda(U(\cdot))$  is strictly increasing and continuous and hence the weakly convergent sequence in Theorem 2 is a Markov chain and, in addition, is mixing [5, Theorem 2]. Thus [4] we cannot have convergence in probability

in Theorem 2.

The quantity converging in law in Theorem 2 has a rather odd appearance since  $U(X_n)$  is present in both numerator and denominator. Tidier versions can be obtained by making estimates of

$$\Lambda(e^{-n}U(X_n))/\Lambda(U(X_n)) = \exp\left\{-Y_n \int_{1-n/Y_n}^1 G(zY_n) dz\right\};$$
 (6)

under appropriate assumptions on the behavior of G. For brevity, we write  $Y_n \equiv \log U(X_n)$  throughout.

Corollary 1. If 
$$p(s) \equiv 0$$
 and  $\lim_{x\to\infty} xG(x) = 0$ , then 
$$\Lambda(e^{-n}U(X_n))/\Lambda(e^n) \stackrel{d}{\to} W$$
,

where W is uniformly distributed on [0,1].

Proof. Since xG(x) + 0, we have

$$\int_{\gamma_n-n}^{\gamma_n} G(y) dy = o \left\{-\log(1-n/\gamma_n)\right\}; \qquad (7)$$

combining Theorem 2, (6) and (7), it follows that, as  $n+\infty$ ,  $n/Y_n \stackrel{d}{\rightarrow} 1$ . Hence, as  $n+\infty$ ,

$$\Lambda(U(X_n))/\Lambda(e^n) = \exp\left\{\int_n^{\gamma_n} G(y) dy\right\}$$

$$= \exp\left\{o(\log[\gamma_n/n])\right\} \stackrel{d}{\leftarrow} 1,$$

and the result follows.

Corollary 2. If 
$$p(s) \equiv 0$$
 and  $\lim_{X\to\infty} xG(x) = a$ ,  $0 < a < \infty$ , then 
$$n^{-1}\log U(X_n) - 1 \stackrel{d}{\to} W$$
, where  $P[W \leq u] = \{u/(1+u)\}^a$ .

Proof. Immediately, from (6), we have

$$\Lambda(e^{-n}U(X_n))/\Lambda(U(X_n)) \overset{a.s.}{\sim} \{1-n/Y_n\}^{a} ,$$

and the result now follows from Theorem 2.

Corollary 3. If  $p(s) \equiv 0$ ,  $\lim_{X\to\infty} xG(x) = \infty$ , and G is regularly varying at infinity, then  $nG(\log U(X_n)) \stackrel{d}{\to} W$ , where W is a standard negative exponential random variable.

Proof. Here, from (6), we have

$$\Lambda(e^{-n}U(X_n))/\Lambda(U(X_n)) = \exp\left\{-Y_n\int_{1-n/Y_n}^1 G(zY_n) dz\right\};(8)$$

combining (8) with Theorem 2 and  $\lim_{x\to\infty} xG(x) = \infty$ , it follows that, as  $n\to\infty$ ,  $n/Y_n\to 0$ . Hence, since G varies regularly,

$$\Lambda(e^{-n}U(X_n))/\Lambda(U(X_n)) \sim \exp\{-nG(Y_n)\}$$
.

and the result follows.

If in Corollary 3 G has a positive index  $\Delta$ , then  $0 < \Delta \le 1$  and the conclusion can be transformed to the form

$$a_n^{-1} \log U(X_n) \stackrel{d}{\leftarrow} W'$$

where W' has the extreme value distribution function  $\exp(-x^{-\Delta})$  and  $G(a_n) = n^{-1}$ .

We now show that for any U the hypotheses of Corollaries 1-3 can be satisfied. Let A be a positive integer valued random variable, define I = [V(A)] and let b be the probability generating function of I. Since

$$V(A) - 1 \le I \le V(A) \tag{9}$$

and log U is SV at infinity, it follows that condition (2) is satis-

fied iff  $E \log^+ A < \infty$ . In [3,§3.1] several examples were given where  $E \log^+ A = \infty$ , and for each of these T(x) = P(A > x) is SV at infinity. We now suppose this to be always the case. It follows then, from Lemma 1 and (9), that  $P(I > x) \sim T(U(x))$  ( $x \rightarrow \infty$ ) and hence an Abelian theorem for power series yields

 $(1-b(s))/(1-s) = \sum s^{j}P(I>j) \sim (1-s)^{-1}T(U(1-s)) \quad (s+1).$  Letting  $s = \exp(-1/V(e^{X}))$  and invoking Lemma 1 once again, we obtain  $G(x) \sim T(e^{X}).$ 

Let  $\log_1 x = \log x$  and  $\log_k x = \log(\log_{k-1} x)$  (k=2,3,...) for all sufficiently large x. In [3] an example was given for which

$$T(x) \sim c \begin{bmatrix} r \\ \pi \\ k=1 \end{bmatrix} \log_{k} x ]^{-1} \qquad (x+\infty)$$

where  $r \ge 2$  and c is a certain constant. Using (10) it is obvious that  $\int_0^\infty G(x) dx = \infty$  and that  $xG(x) \to 0$   $(x \to \infty)$  and hence this example satisfies the conditions of Corollary 1.

Discrete distributions were also constructed in [3] for which

$$T(x) \sim a/\log x$$
  $(0 < a < \infty)$ ;

$$T(x) \sim c(\log x)^{-(\delta-1)}$$
 (0 < c < \infty, 1 < \delta < 2);

and

$$T(x) \sim (c/b)(\log_{r} x)^{-b}$$
  $(0 < b, c < \infty, r \ge 2).$ 

Using (10) we see that these examples satisfy the hypotheses of Corollaries 2 and 3.

Finally, we show that the neatest version of Theorem 2 that could be hoped for is in fact impossible.

Theorem 3. There is no sequence of constants  $(c_n)$  such that  $c_n^{-1}U(X_n)$  has a limit in distribution which is neither defective nor degenerate at zero.

<u>Proof.</u> For any  $0 < x_1 < x_2 < \infty$ , let  $p_n(x_1, x_2) = P[x_1 \le c_n^{-1}U(X_n) \le x_2]$ . Then, since  $\Lambda(e^{-n}U(\cdot))/\Lambda(U(\cdot))$  is continuous and strictly increasing, it follows also that

 $p_{n}(x_{1},x_{2}) = P[r_{n}(x_{1}) \leq \Lambda(e^{-n}U(X_{n}))/\Lambda(U(X_{n})) \leq r_{n}(x_{2})] ,$  where

 $0 < r_n(x) \equiv \Lambda(e^{-n}c_nx)/\Lambda(c_nx) < 1.$  If  $(c_n)$  is such that  $c_n^{-1}U(X_n)$  is to converge in distribution, we must have  $c_ne^{-n} \to \infty$  because of Theorem 1. Hence, since  $\Lambda$  is SV, we see that  $r_n(x_2) = r_n(x_1)[1+o(1)]$  as  $n\to\infty$ . Choose any subsequence  $(n_k)$  such that  $r_n(x_1) + r$  for some  $r\in[0,1]$ . Then, for any  $\epsilon > 0$ , the intervals  $\left\{r_n(x_1), r_n(x_2)\right\}$  belong to  $(r-\epsilon, r+\epsilon)$  for all k sufficiently large. It follows from Theorem 2 that  $p_n(x_1, x_2) \to 0$ , and hence that, if  $c_n^{-1}U(X_n)$  converges in distribution, its limit puts no mass on  $(0, \infty)$ .

It is interesting to note the contrast between the cases p(x) > 0 and  $p(x) \equiv 0$ . In the former, the asymptotic behavior is dominated by the underlying Galton-Watson process, and the effect of immigration, apart from preventing extinction, is seen only in the distribution of the limit of  $U(X_n)e^{-n}$ : eventually, the contribution of the immigration process becomes negligible. However, when  $p(x) \equiv 0$ , the immigration distribution has such a broad tail that  $U(X_n)e^{-n}$  is pushed off to infinity a.s. by the infinite sequence of occasional, but very large, inflows of immigrants. The character of Theorem 2 and its Corollaries, giving limits in distribution but not with probability one, reflects the nature of the immigration process rather than that of the Galton-Watson process. In particular, unlike the case

when p(x) > 0, the limiting distribution appearing in Theorem 2 is the same, whether or not the Galton-Watson process is regular or irregular.

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If  $s = \infty$ . When  $s = \infty$  limit theorems are found for  $(U(X_n))$  which

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